



BS/BSC

in Applied AI and Data Science

Foundations of Statistics & Probability

IIT Jodhpur



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Foundations of Statistics & Probability

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Module 02

Measures of Central Tendency

Exploring data variability



Recap

Data organization and visualization

Introduction to Descriptive Statistics

Mean, median, and mode

Population vs. Sample Variance



In this video,

Chebychev's Identity

Apply Chebyshev's Inequality to datasets with unknown distributions.

Interpret the probability bounds for real-world scenarios



Standard Deviation & Probability



Standard Deviation & Probability

The notation $P(X > a)$ represents the **probability that the random variable X takes a value greater than a .**

- X is a **random variable**, meaning it represents different possible outcomes of a process (e.g., dice roll, test scores, AI response time).
- $P(X > a)$ means we are calculating the probability that X will be greater than a specific value a .
- It is equivalent to asking:
“**What is the chance that X will be more than a ?**”

Chebychev's Identity

Chebyshev's Inequality is a fundamental theorem in probability that provides a bound on how much of the data deviates from the mean.

Chebyshev's Inequality states that **for any dataset** (not necessarily normal),

$$P(|X - \mu| \geq k\sigma) \leq \frac{1}{k^2}$$

X = A random variable (e.g., test scores, stock prices, etc.)

μ = Mean (average value of the dataset)

σ = Standard deviation (spread of data)

k = Number of standard deviations away from the mean

The probability that a value is at least k standard deviations away from the mean is at most $1/k^2$.

Why should you know about Chebychev's Inequality?

-  It does not assume normality, making it broadly applicable.
-  It gives a worst-case bound, meaning it guarantees that extreme values do not occur too often (helps in setting safety bounds)
-  It is widely used in anomaly detection, AI model evaluation, and risk assessment.

Example Fraud Detection in AI Systems

A company tracks daily transactions of an e-commerce AI system.

The mean transaction value is:

$$\mu = 200 \text{ dollars}$$

$$\sigma = 50 \text{ dollars}$$

Fraudulent transactions usually have extremely high or low values.

E.g. How often a transaction will be more than 3 standard deviations away from the mean?

AI fraud detection models can use this to set risk thresholds.

At most 11.11% of transactions will be below 50 dollars or above 350 dollars.



$$P(|X - 200| \geq 3(50)) \leq \frac{1}{3^2} = \frac{1}{9} \approx 11.11\%$$

Scenario AI-powered Delivery System





Scenario AI-powered Delivery System

An AI-driven delivery system predicts estimated delivery times (in minutes) for a food delivery app.

After analyzing data from **thousands of orders**, you find:

The mean delivery time (μ) is **30 minutes**.

The standard deviation (σ) is **5 minutes**.

Scenario AI-powered Delivery System

The mean delivery time (μ) is 30 minutes.

The standard deviation (σ) is 5 minutes.



Your company wants to guarantee customers that deliveries will be within **20 minutes to 40 minutes most of the time**.

Use Chebyshev's inequality to determine the maximum proportion of deliveries that might fall **outside** this range.



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Use Chebyshev's inequality to determine the **maximum proportion** of deliveries that might fall **outside** this range.

- The number of standard deviations away from the mean:

$$k = \frac{40 - 30}{5} = 2$$

- Using Chebyshev's inequality:

$$P(|X - 30| \geq 2\sigma) \leq \frac{1}{2^2} = \frac{1}{4} = 0.25$$

Scenario AI-powered Delivery System

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Use Chebyshev's inequality to determine the **maximum proportion** of deliveries that might fall **outside** this range.

$$P(|X - 30| \geq 2\sigma) \leq \frac{1}{2^2} = \frac{1}{4} = 0.25$$

Ans: At most **25%** of deliveries might take **less than 20 minutes or more than 40 minutes.**

Scenario AI-powered Delivery System

The mean delivery time (μ) is 30 minutes.

The standard deviation (σ) is 5 minutes.

The company considers a delivery "very late" if it takes more than **3 standard deviations from the mean**.

Use Chebyshev's inequality to find out: at most, what **percentage** of deliveries will take more than 45 minutes?



Scenario AI-powered Delivery System

The mean delivery time (μ) is 30 minutes.

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At most, what percentage of deliveries will take more than 45 minutes?

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The company considers a delivery "very late" if it takes more than **3 standard deviations from the mean**.

Use Chebyshev's inequality to find out:

At most, what percentage of deliveries will take more than 45 minutes?

Using Chebyshev's inequality:

$$P(|X - 30| \geq 3\sigma) \leq \frac{1}{3^2} = \frac{1}{9} \approx 0.1111$$

At most **11.11%** of deliveries might take more than 45 minutes.



Summary

Chebyshev's Inequality is a powerful tool when data distributions are unknown.

It provides **minimum guaranteed probability bounds** for datasets.

Coming up next...

Normal distributions and paired datasets